

This work was supported in part by the Spanish Ministry of Economy and Competitiveness under TestEAMoS (TIN2016-76956-C3-1-R) project and ERDF funds.

Test Driven Anonymization Technique for Artificial Intelligence

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Netflix Contest Dataset (I)



Dataset

			★★★★★ Th 14 June	
	★★★★★ Th 14 June			★★★★★ Sat 3 January
		★★★★★ Mon 16 July		
			★★★★★ Sat 12 September	
	★★★★★ Wed 18 July		★★★★★ Fri 11 October	★★★★★ Mon 22 January
		★★★★★ Wed 22 March		
		★★★★★ Fri 25 May		

- Release an useful dataset without identifiers to develop recommenders with it

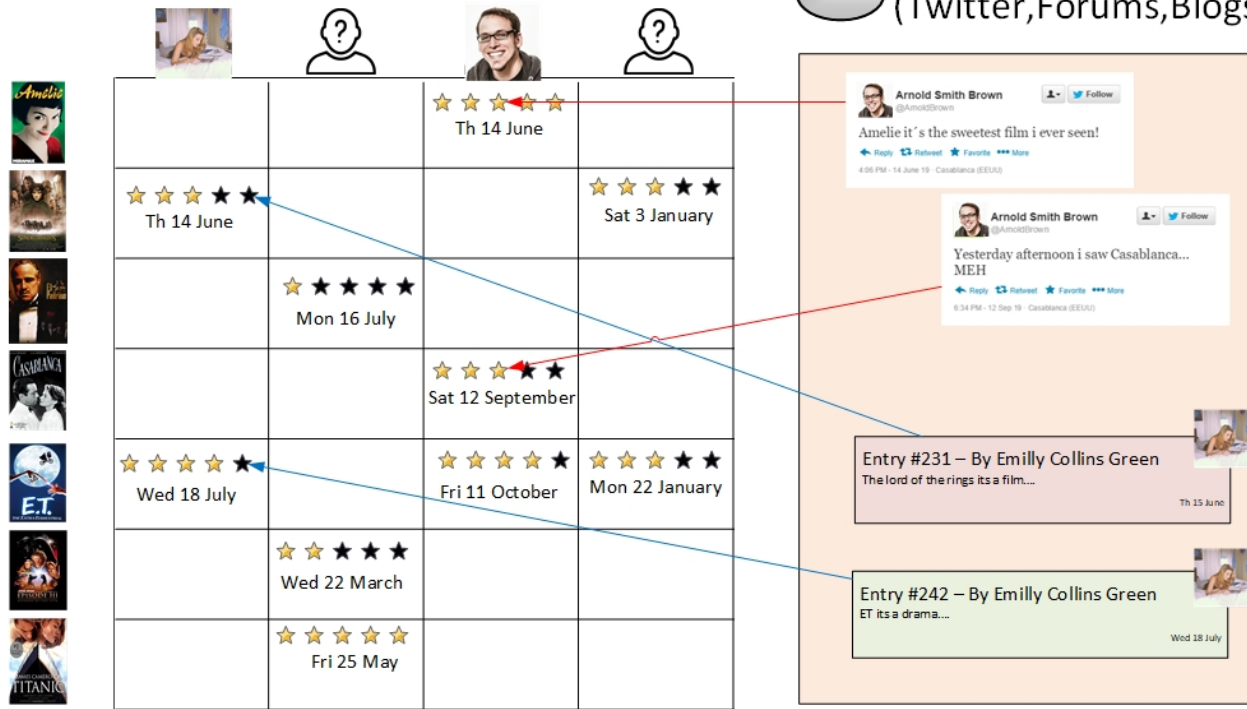
Netflix Contest Dataset (II)



Dataset

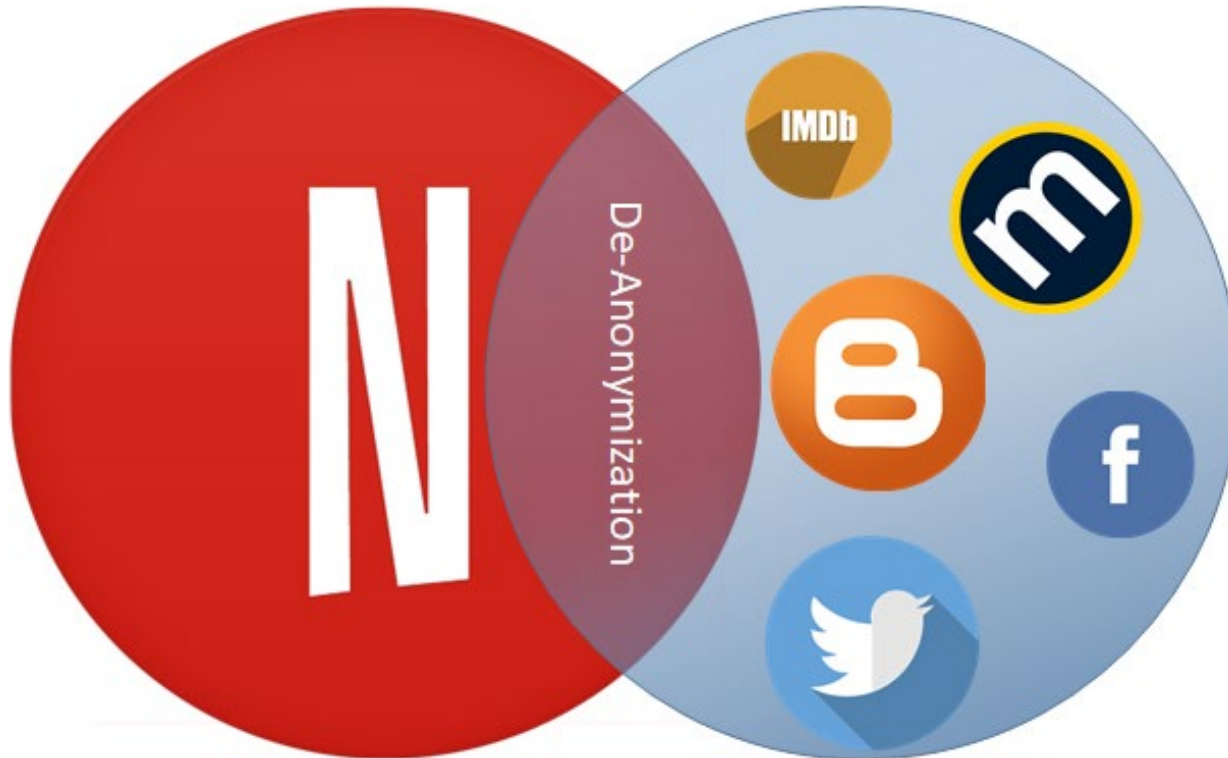


Public Data
(Twitter, Forums, Blogs...)



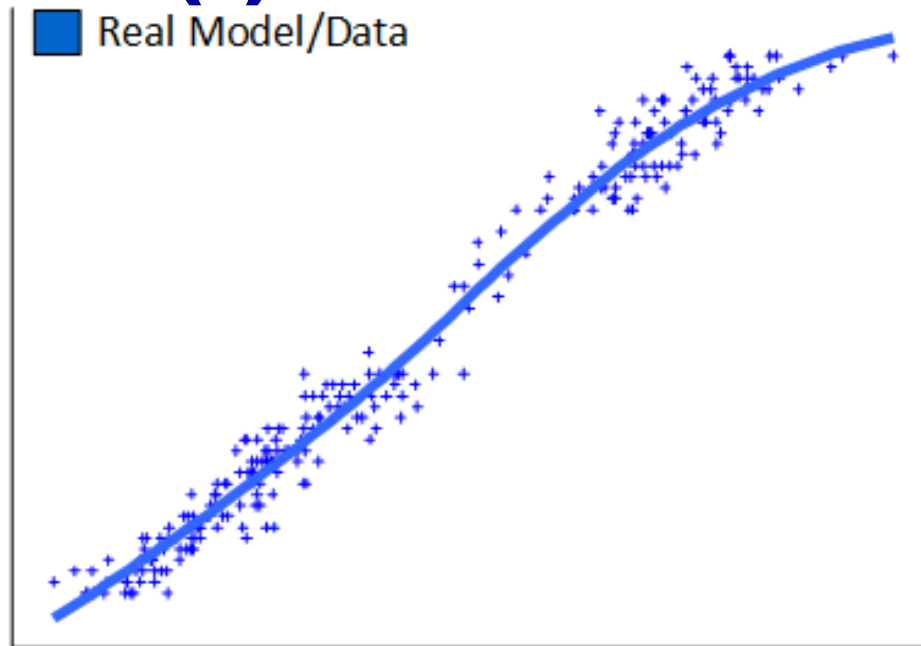
- The information shared by Netflix with another external data like Tweets or Blogs Post allow to identify users

Netflix Contest Dataset (III)



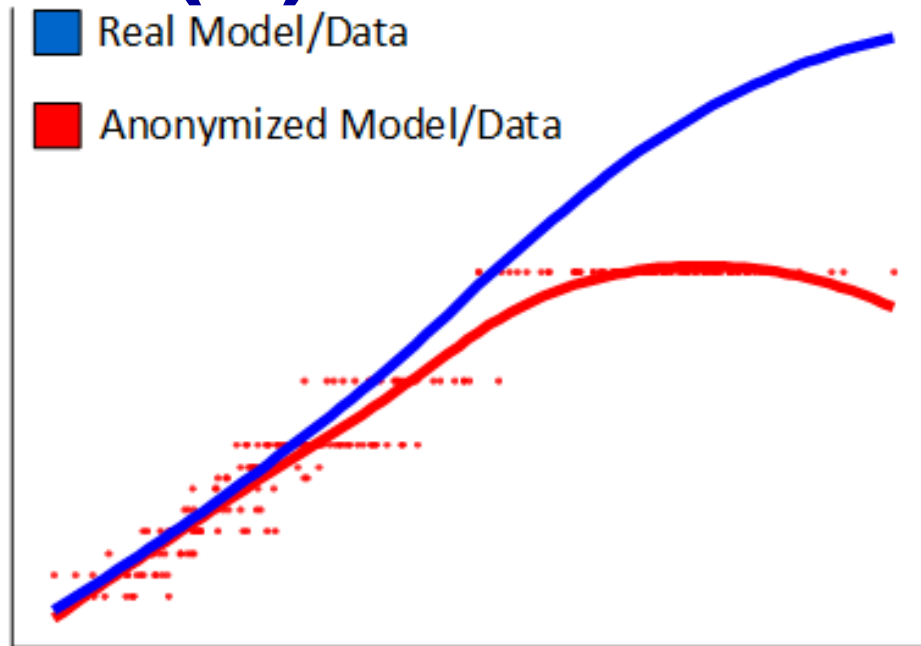
- Cross information of some public sources

Context (I)



- Anonymization effect
 - Improve individuals privacy also loss of information
 - Alters data and affects developments highly dependent of the data (i.e Artificial Intelligence models)

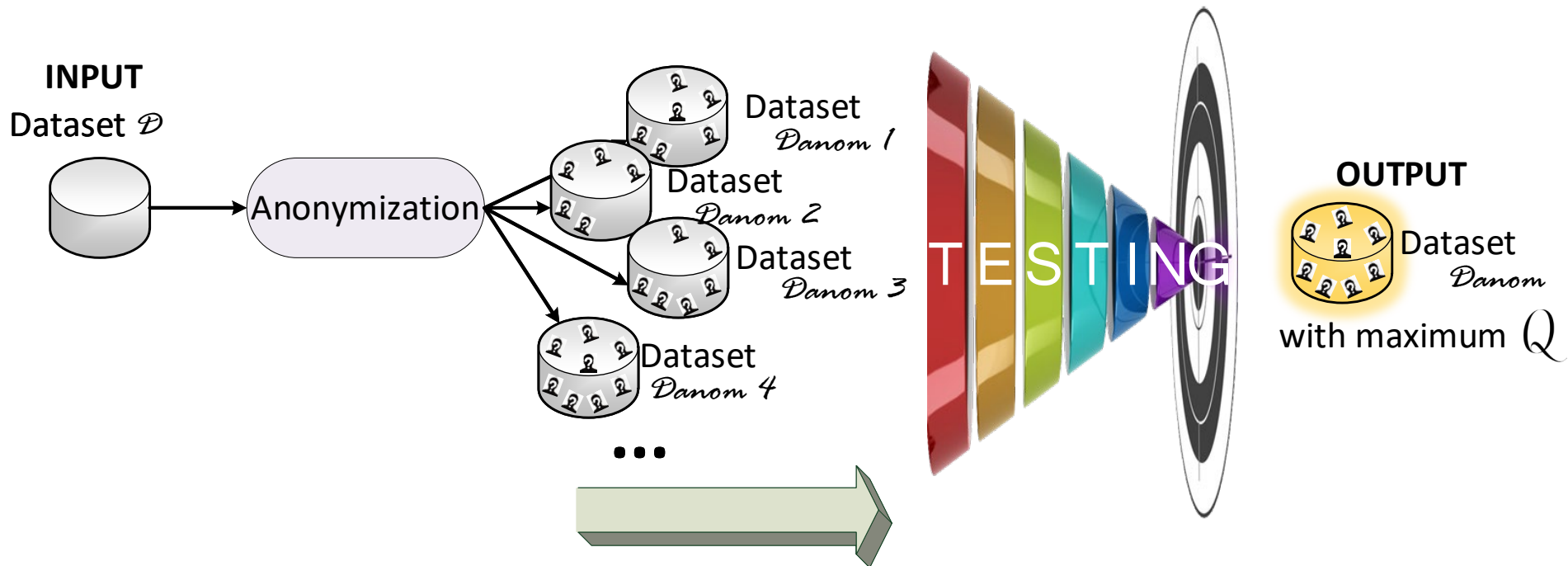
Context (II)



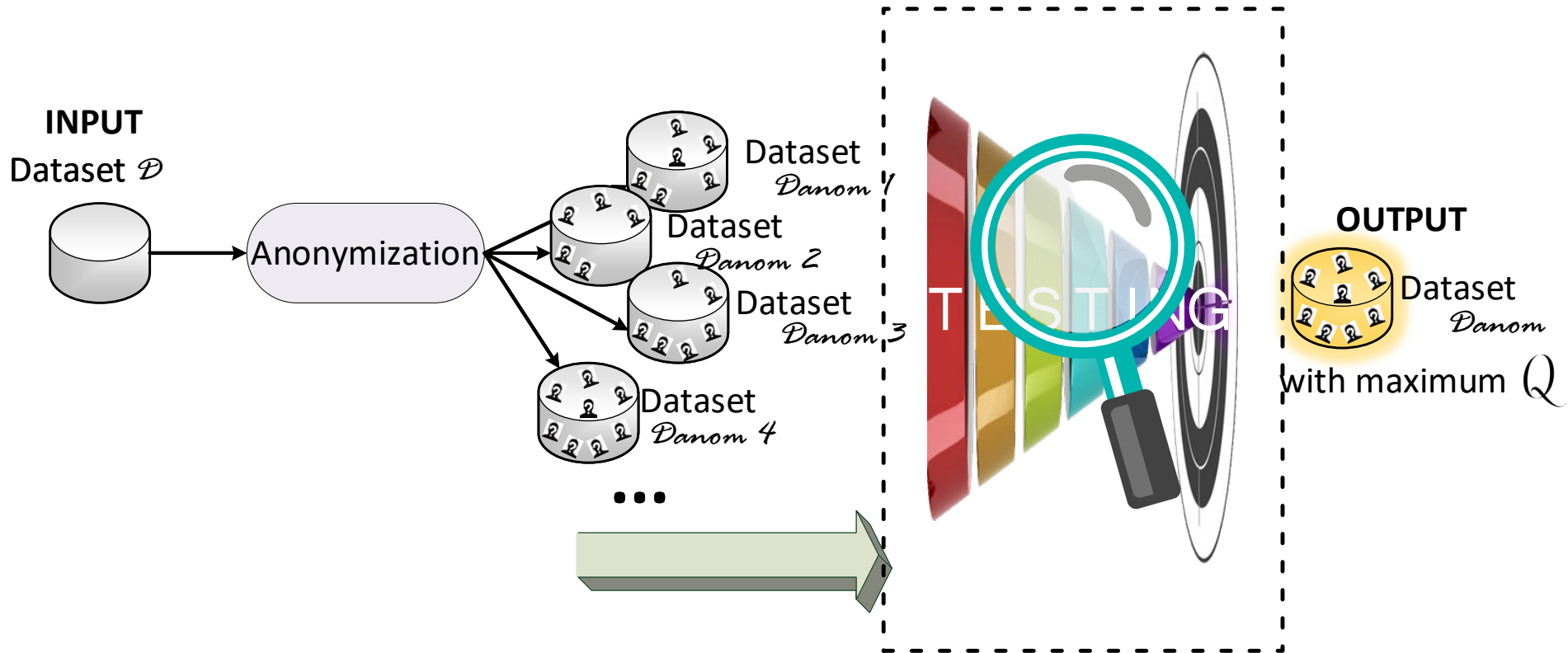
■ Fault Masking

- May be the case that an AI Model that works well in the development stage fails dramatically during production

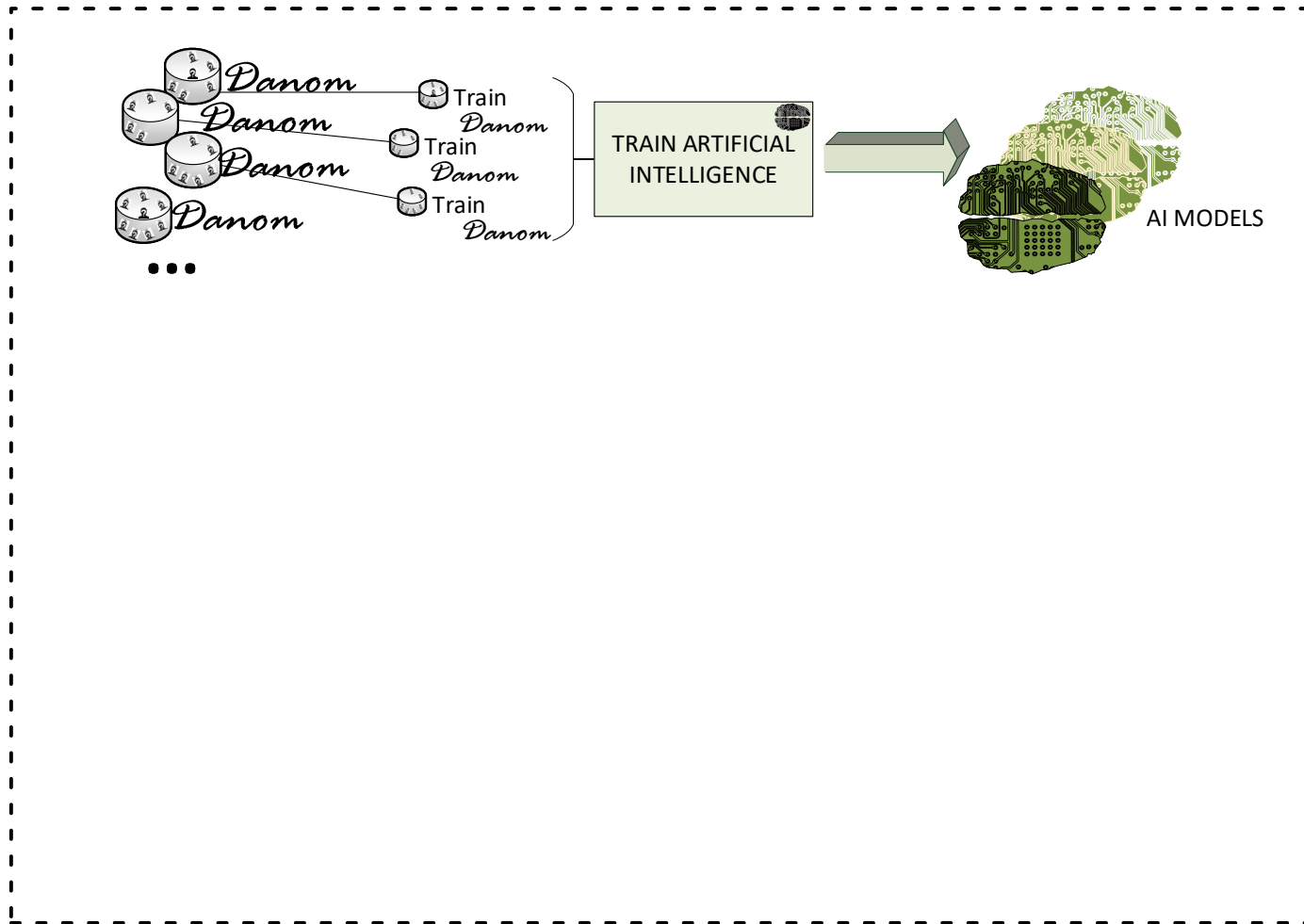
Test Driven Anonymization Overview



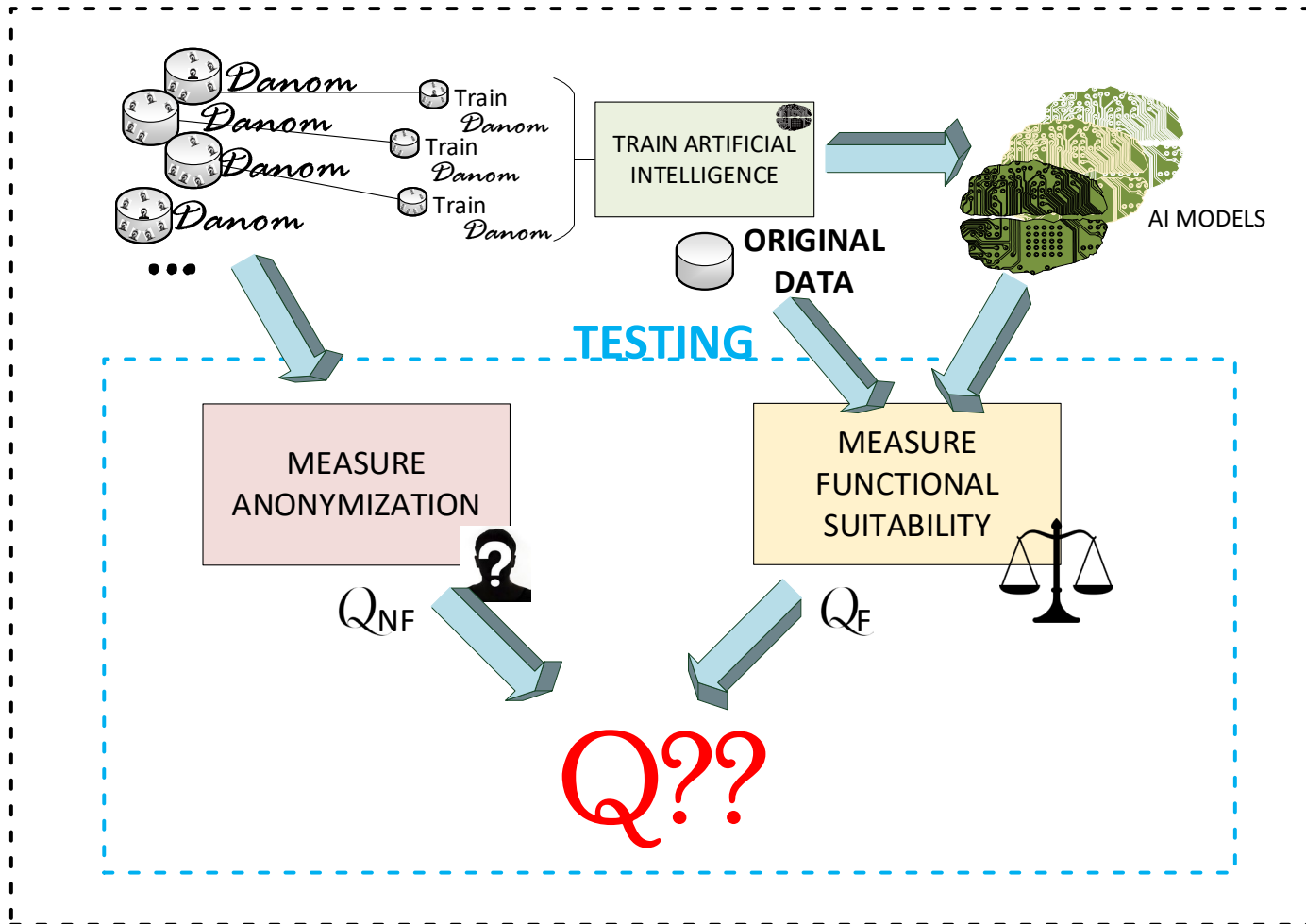
Test Driven Anonymization Overview



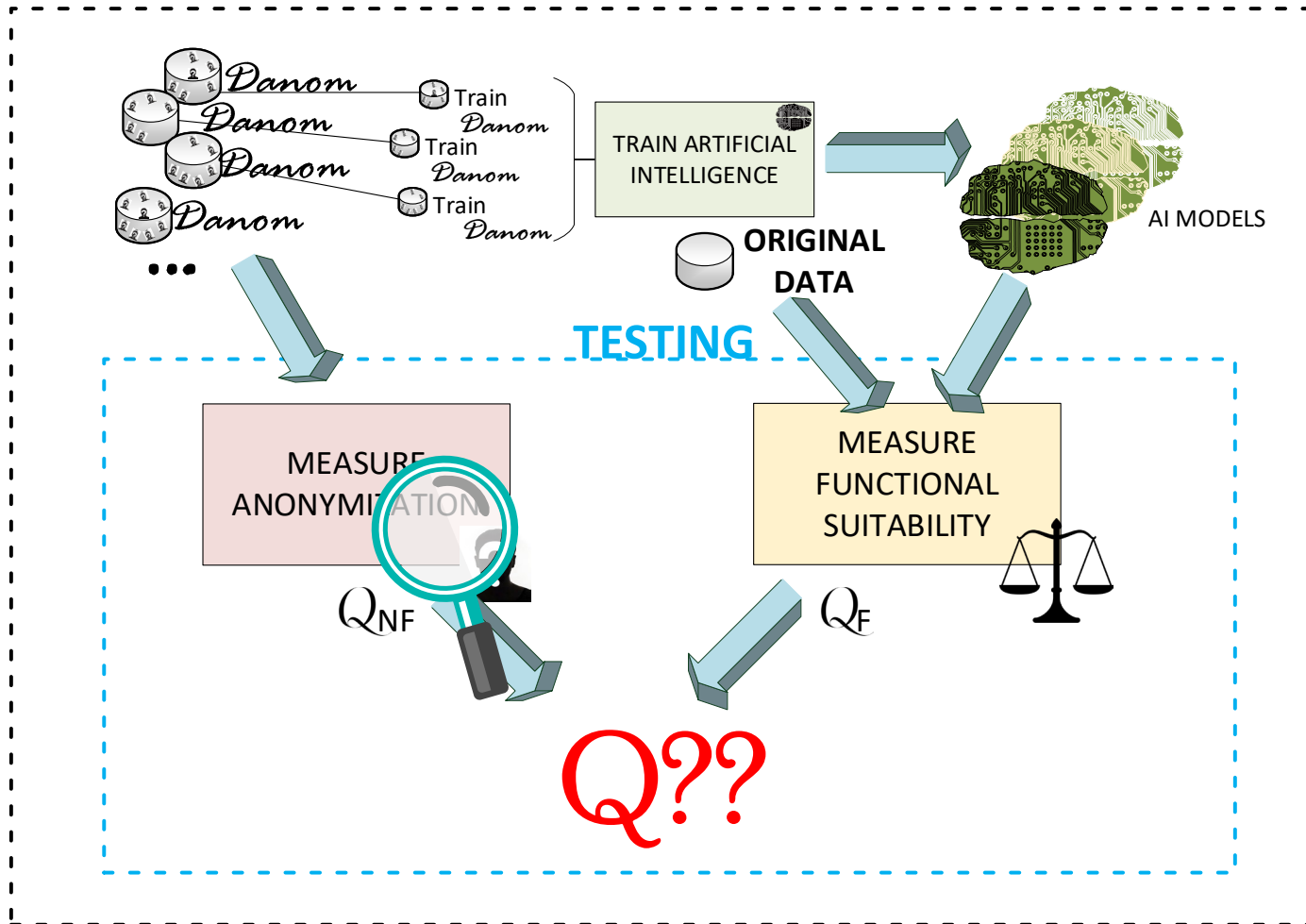
Test Driven Anonymization Approach



Test Driven Anonymization Approach

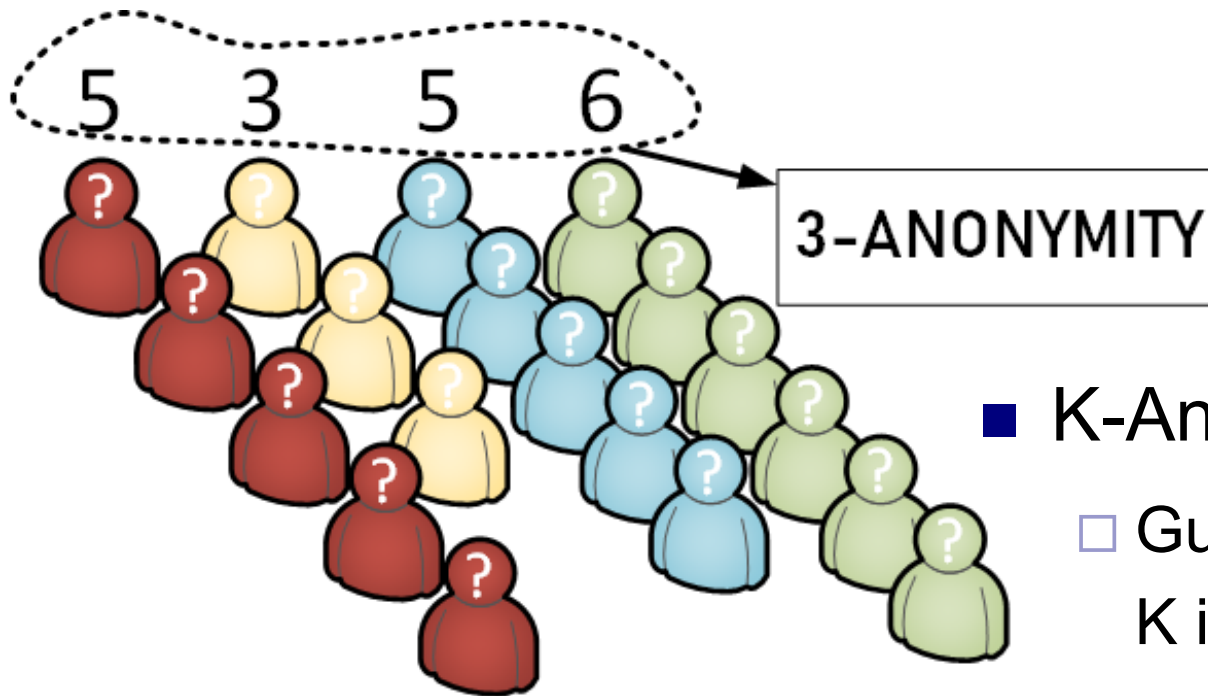


Test Driven Anonymization Approach



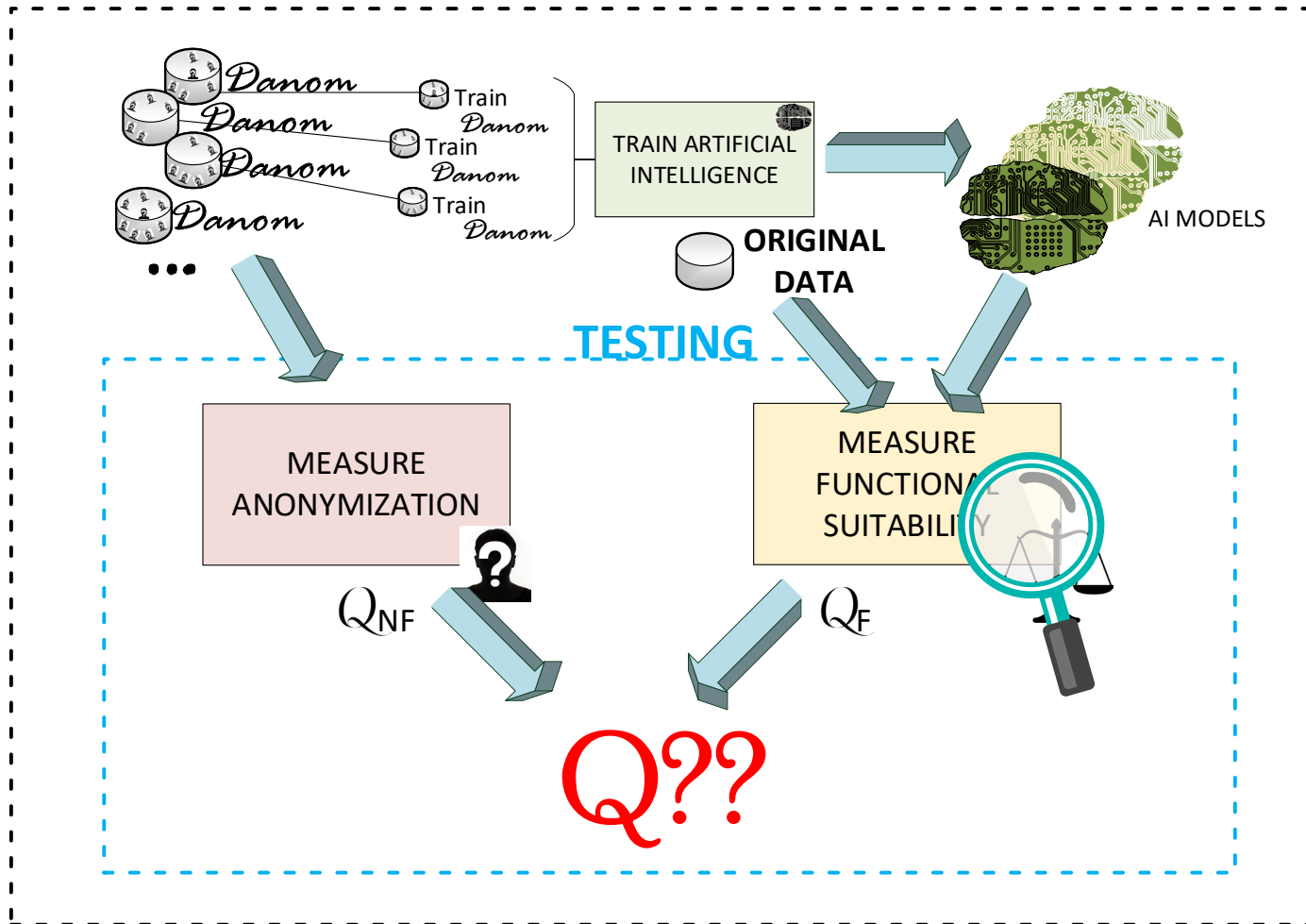
Test Driven Anonymization

Non-Functional Quality



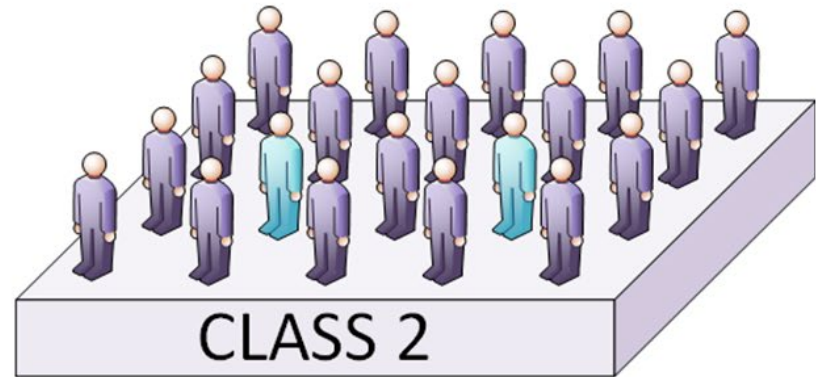
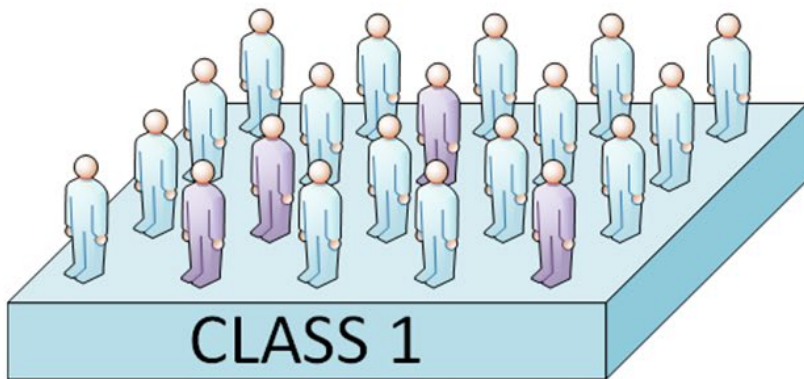
- K-Anonymity:
 - Guarantees more than K individuals for each combination of pseudo-identifiers

Test Driven Anonymization Approach



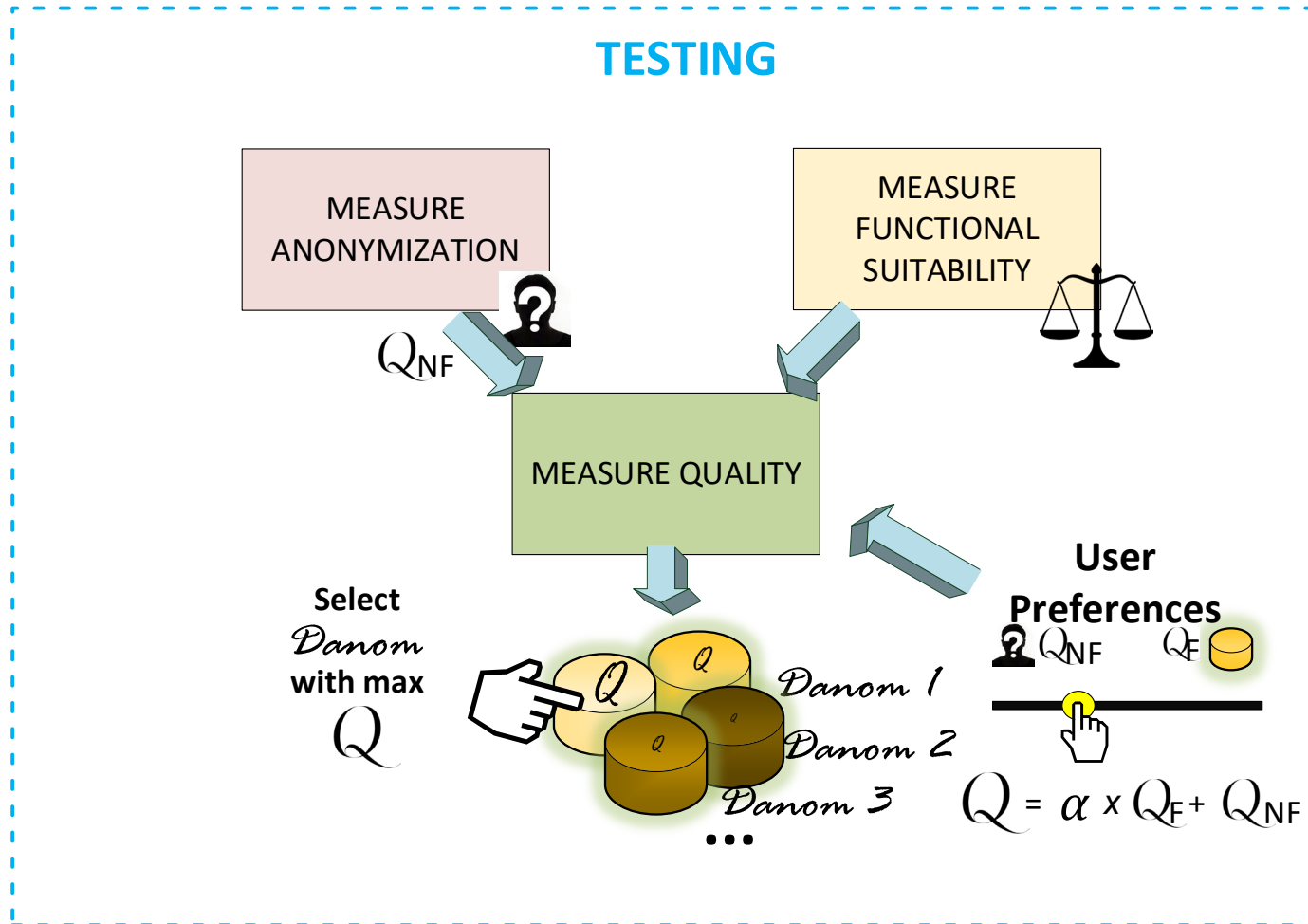
Test Driven Anonymization

Functional Suitability



$$Accur = \frac{N^a \text{ (with 2 checkmarks)} + N^a \text{ (with 1 checkmark)}}{N^a \text{ (with 1 figure)}}$$

Test Driven Anonymization Quality Metric

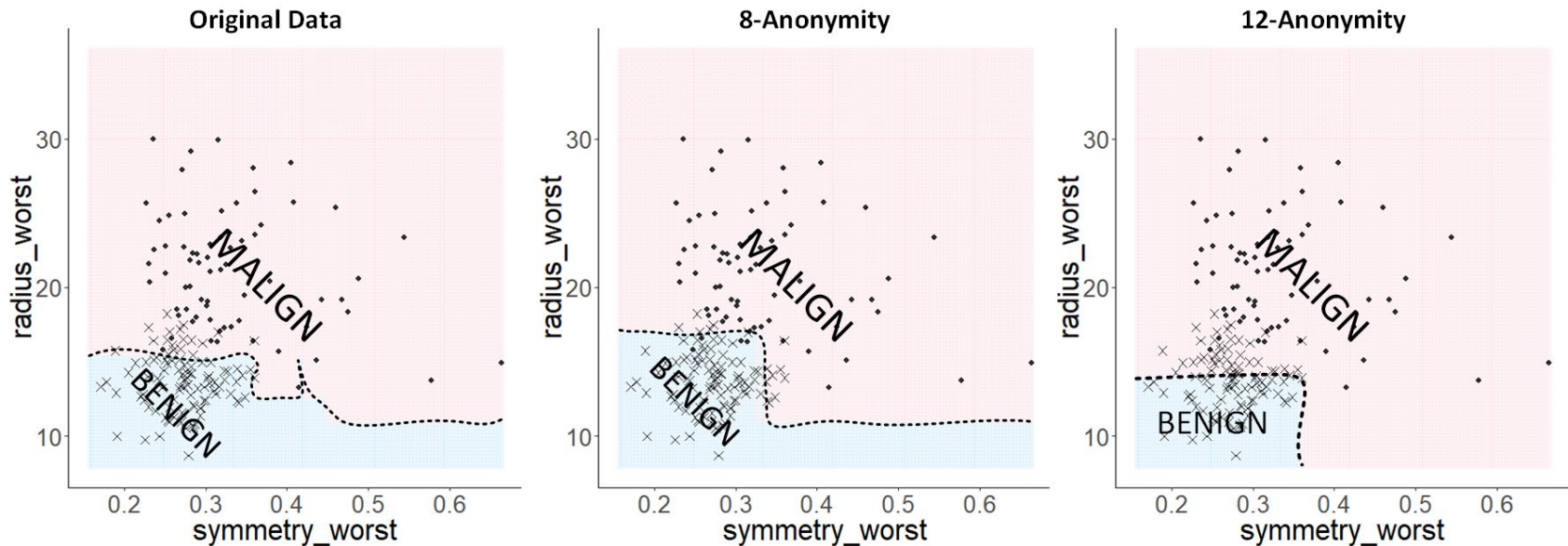


Experimentation (I)

id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	...
1	M	17.99	10.38	122.8	1001	...
2	M	20.57	17.77	132.9	1326	...
3	M	19.69	21.25	130	1203	...
4	M	11.42	20.38	77.58	386.1	...
5	M	20.29	14.34	135.1	1297	...
6	M	12.45	15.7	82.57	477.1	...
7	M	18.25	19.98	119.6	1040	...
8	M	13.71	20.83	90.2	577.9	...
9	M	13	21.82	87.5	519.8	...
10	M	12.46	24.04	83.97	475.9	...
11	M	16.02	23.24	102.7	797.8	...
12	M	15.78	17.89	103.6	781	...
13	M	19.17	24.8	132.4	1123	...
14	M	15.85	23.95	103.7	782.7	...
15	M	13.73	22.61	93.6	578.3	...

- Two public domain datasets related with medical insurance and health:
 - Breast-Cancer(Wisconsin) Dataset
 - Medical Cost Personal Dataset
- Using a generalization method with numerical pseudo-identifiers

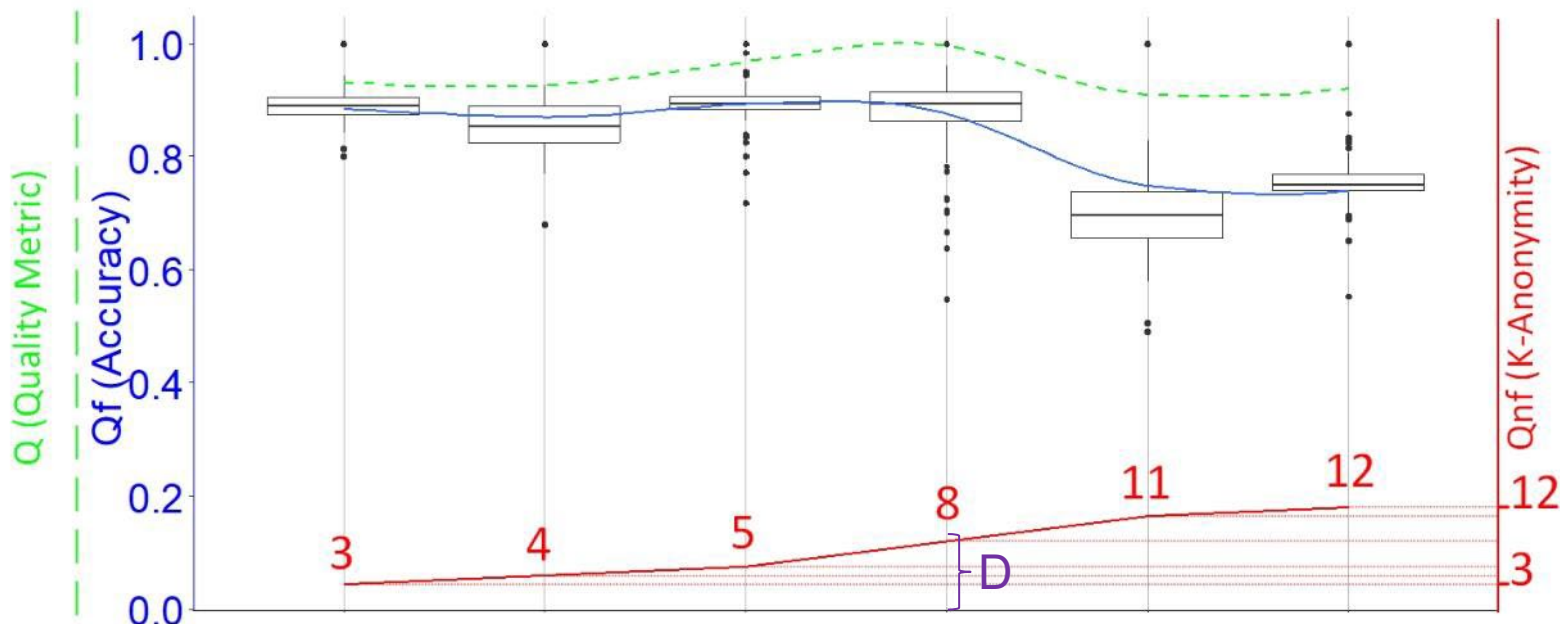
Experimentation (II)



■ Wisconsin Dataset:

- Uses a Random-Forest AI Algorithm for classification, with two size tumor metrics of input
- Try to predict if the tumor is malign or benign

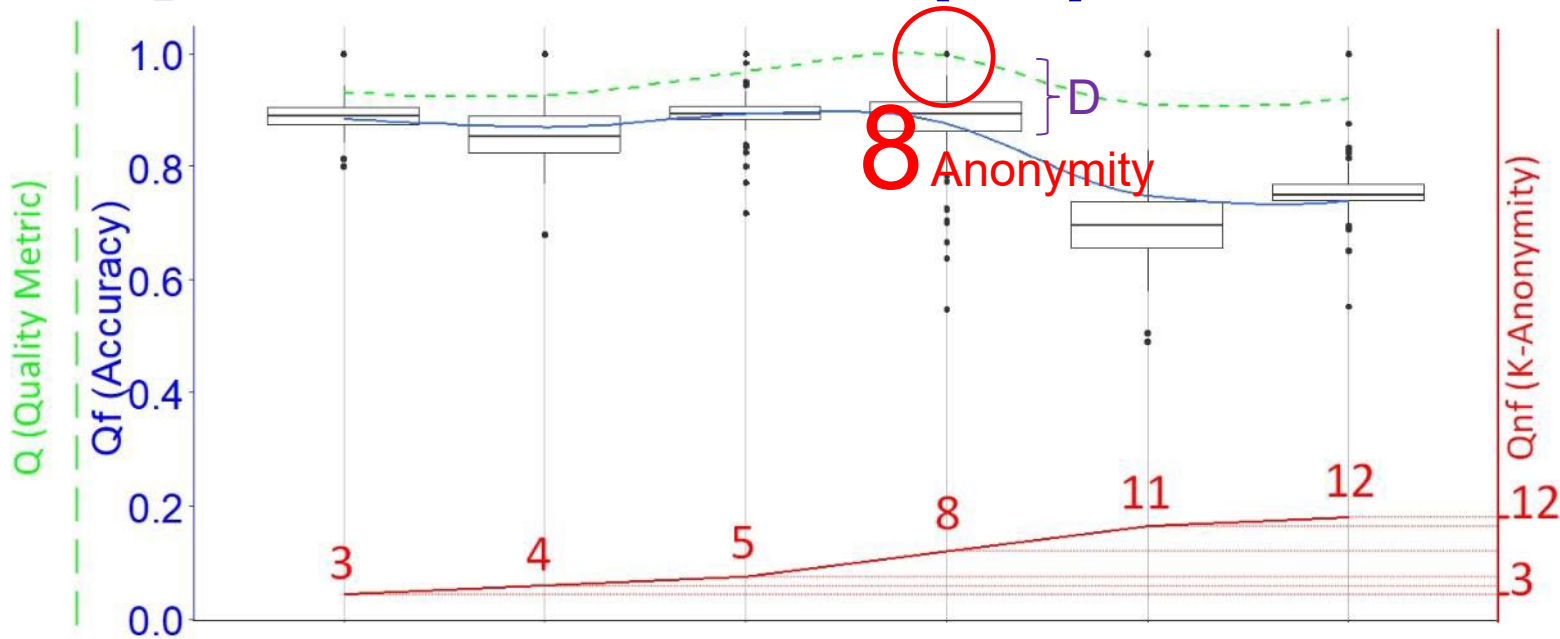
Experimentation (III)



■ Wisconsin Dataset:

- Reach the trade-off between functional and non functional quality with a 8-Anonymization

Experimentation (IV)



■ Wisconsin Dataset:

- Reach the trade-off between functional and non functional quality with a 8-Anonymization

Conclusions and future work

■ Conclusions

- TDA achieves a trade off between privacy and functional suitability.
- TDA allows provider the releasing of useful datasets for developing AI tools

■ Future Work

- Evaluation in more case studies and techniques
- Automation the approach
- Evaluate the dependence between TDA and AI algorithms

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Any Question?

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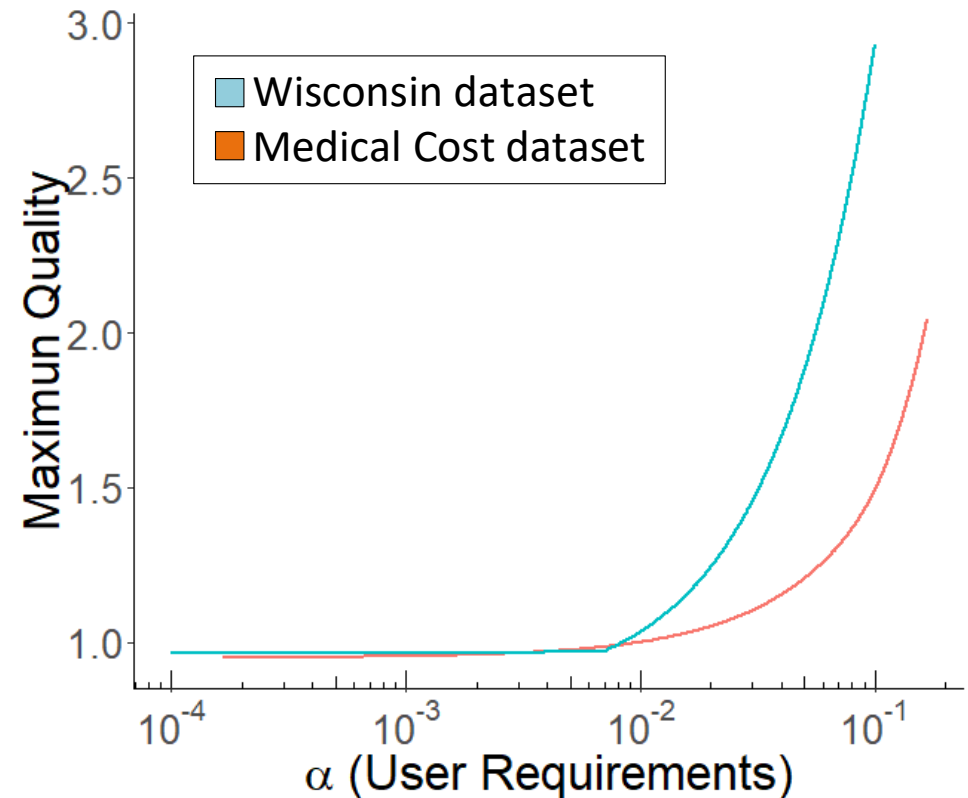


User Preferences

- Represented by α value
 - Prior data utility or data privacy.
- Our quality metric :
 - $Q = QF + \alpha \cdot QNF$

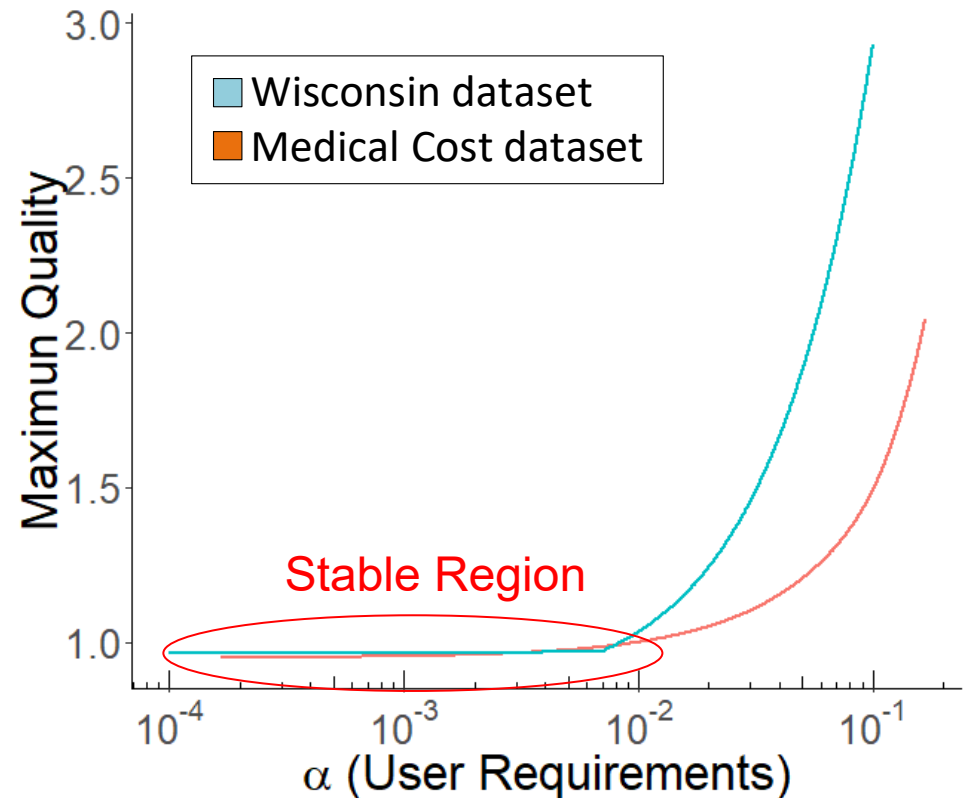
User Requirements

- Selecting the value:



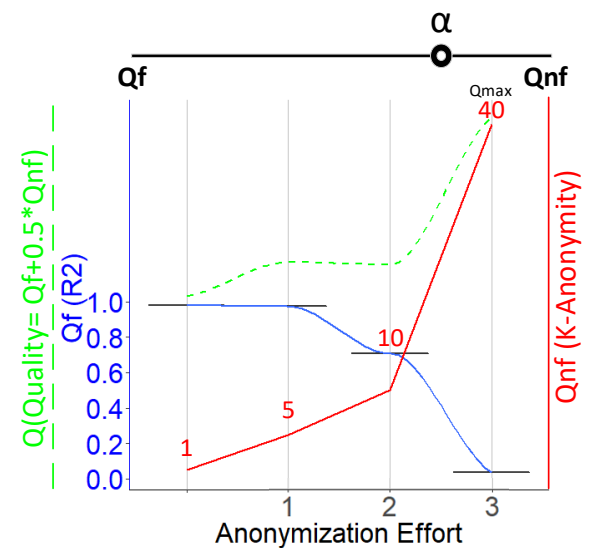
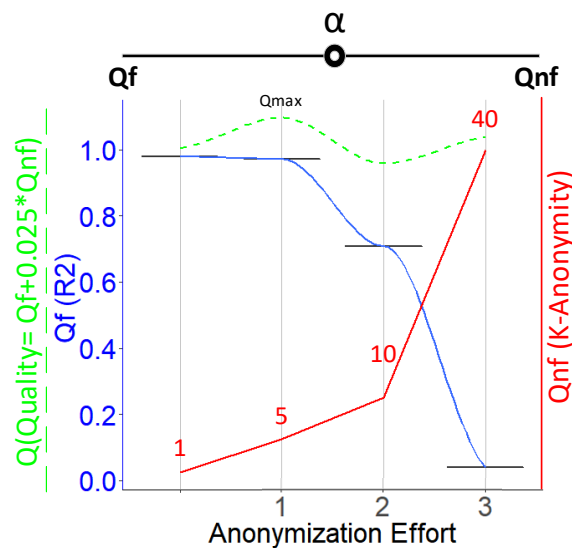
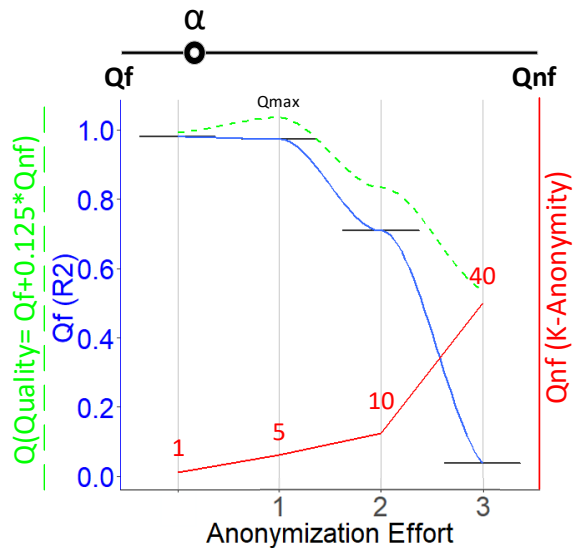
User Requirements

- Selecting the value:



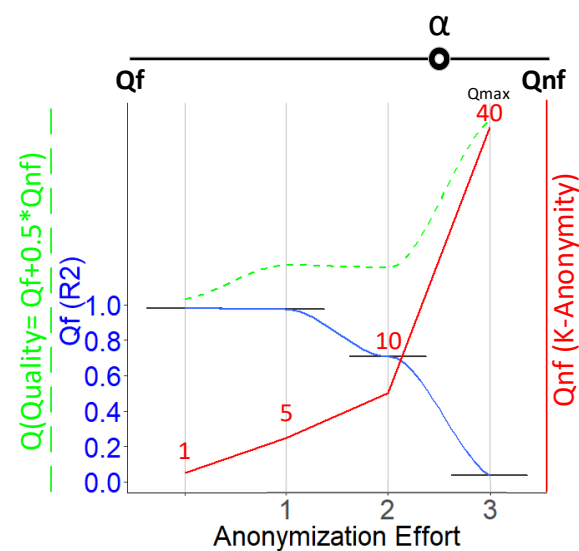
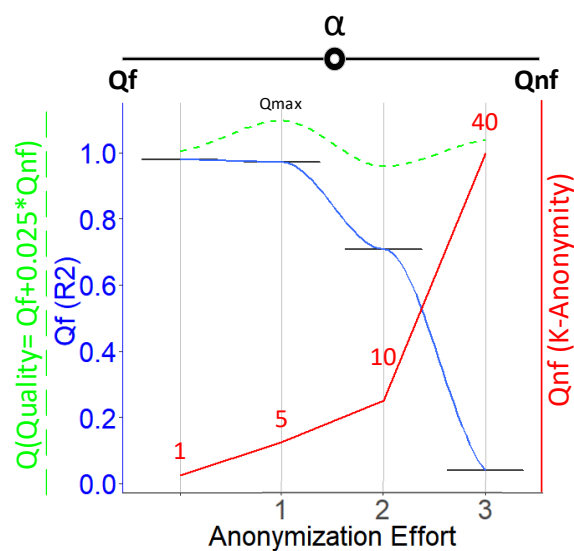
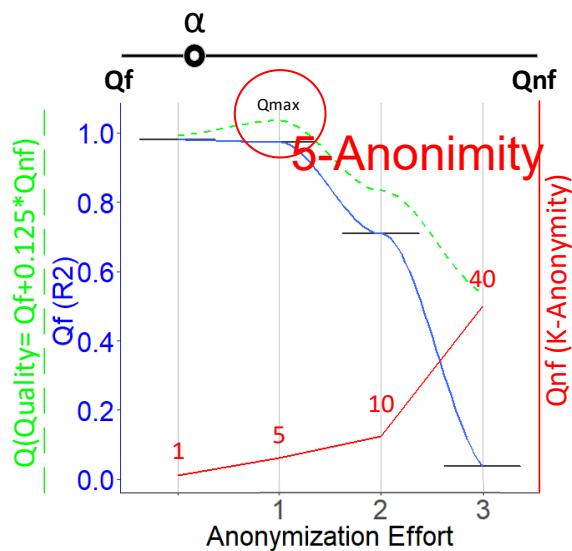
User Preferences

■ Different preferences:



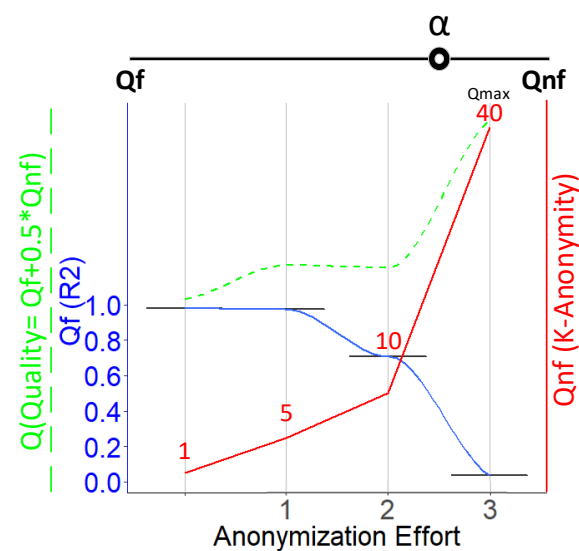
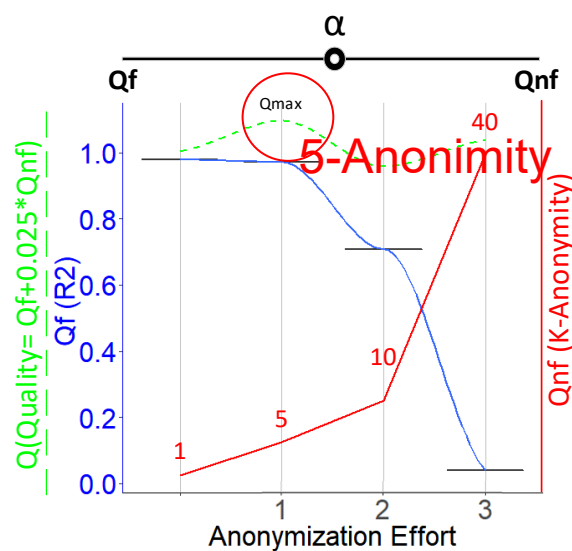
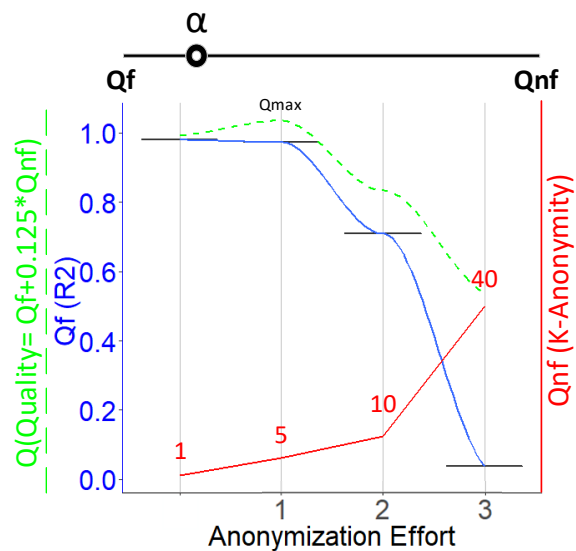
User Preferences

■ Different preferences:



User Preferences

■ Different preferences:



User Preferences

- Different preferences:

